# Enhancing Surface Data Assimilation and Near-Surface Weather Forecasts in NGGPS through Improved Coupling between the Land Surface and Atmosphere

#### Principal Investigator

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#### **Outline**

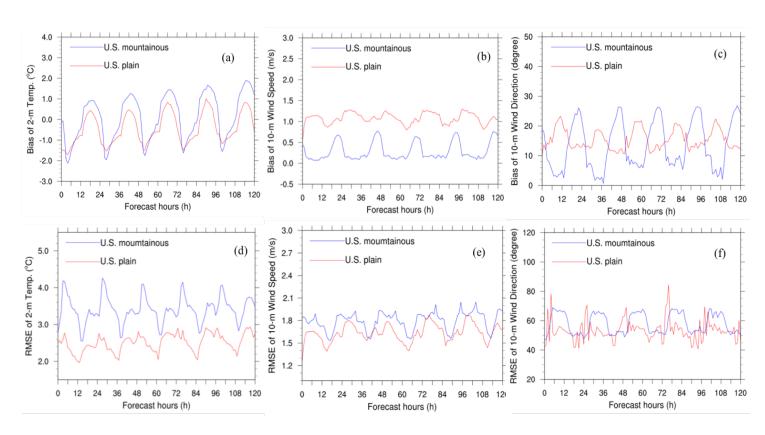
#### **Summary of the progress**

- Evaluating near-surface weather forecast errors
- Understanding covariances between soil and atmospheric states
  - Observational analysis
  - A single column model study
  - A strongly coupled land-atmospheric system (e.g., WRF-Noah)
- Examining the influence of strong coupling on soil moisture data assimilation with SMAP satellite data
- On-going development
  - Strong coupling within the GSI framework
  - NGGPS (NCEP FV3) coupling with NASA LIS

#### **Evaluating near-surface weather forecast errors**

#### Mean bias and RMSE for 2-m temperature and 10-m winds

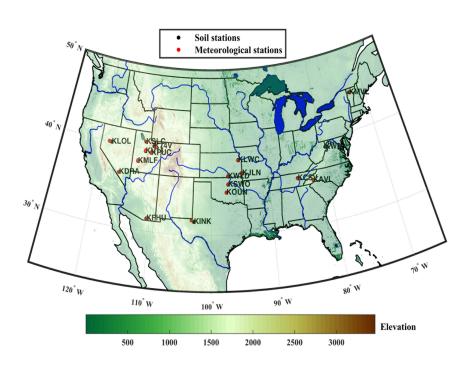
GFS. - U. S. Mountainous vs. U. S. Plains 00UTC FCST, June 2016



### Understanding covariances between soil and atmospheric states Observational analysis (Student J. Liu)

#### Is soil moisture a major factor that affects near-surface weather forecasts?

- The meteorological observations, soil moisture data, and soundings from surface Mesonet, Climate Reference Network (CRN), Soil Climate Analysis Network (SCAN) network, and University of Wyoming sounding databases
  - The correlation coefficient (R)
  - Information flow analysis (Liang, 2014 and 2015)



### 16 soil moisture, 16 meteorological stations, 2 sounding stations (2008-2016)

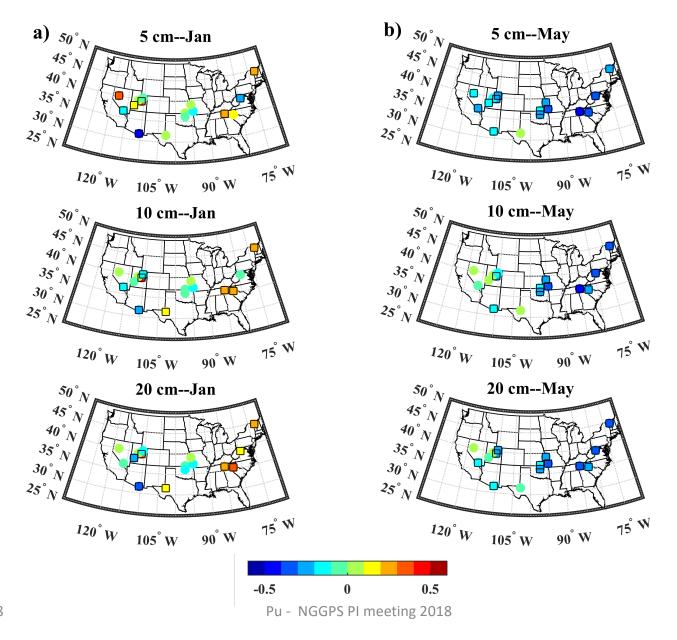
KPUC, KU14, KMLG, and K74V: mountain area with shrubland and grassland. KWLD, KSWO, KJLN, and KLWC: plain area with grassland.

KINK, KFHU, KDRA and KLOL: desert area with shrubland.

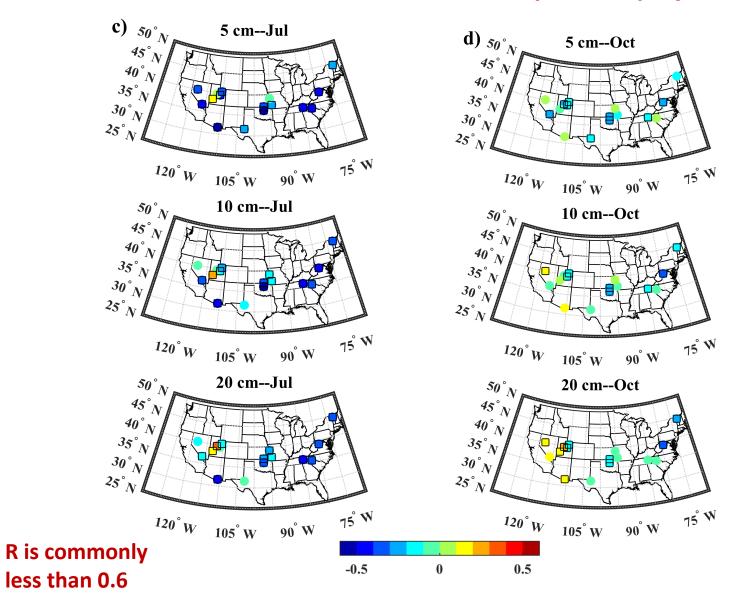
KW99, KAVL, KCSV and KMVL: mountain area with forest.

2 sounding stations: KSLC (Intermountain West) and KOUN (Great Plain)

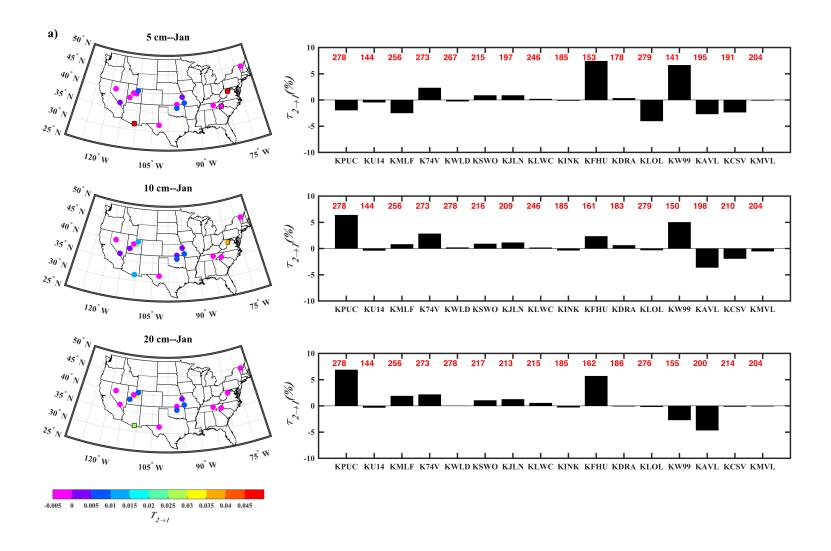
#### **Correlations between soil moisture and 2-m temperature (January and May)**



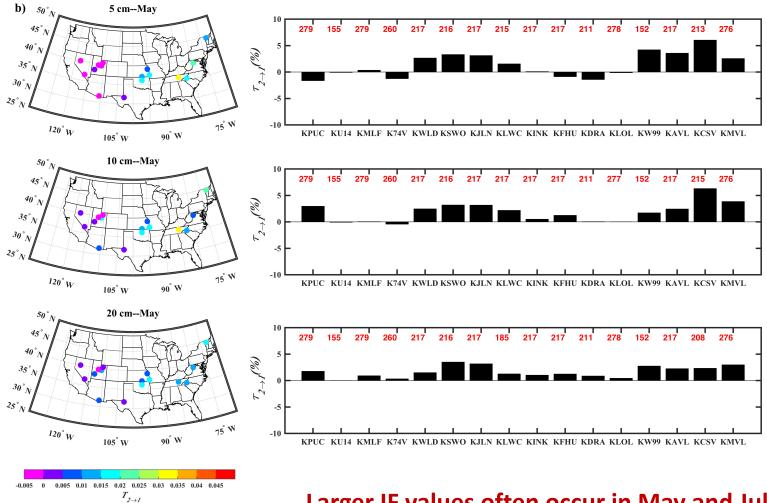
#### **Correlations between soil moisture and 2-m temperature (July and October)**



#### The information flows from soil moisture to 2-m temperature (January)



#### The information flows (IF) from soil moisture to 2-m temperature (May)



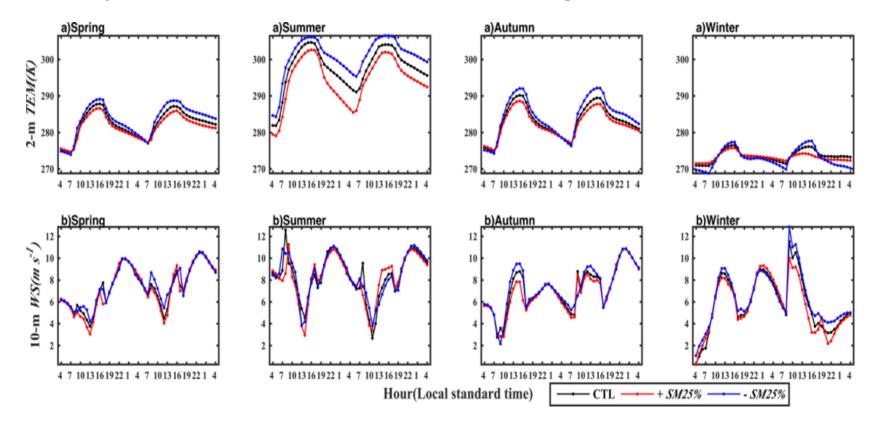
Larger IF values often occur in May and July

#### Understanding covariances between soil and atmospheric states

A single column model study (Student J. Liu)

- WRF single column model; WRF Version 3.8.1
- RRTM longwave radiation/ Dudhia shortwave radiation/ Noah Land Surface model / YSU PBL / WSM-6 microphysics

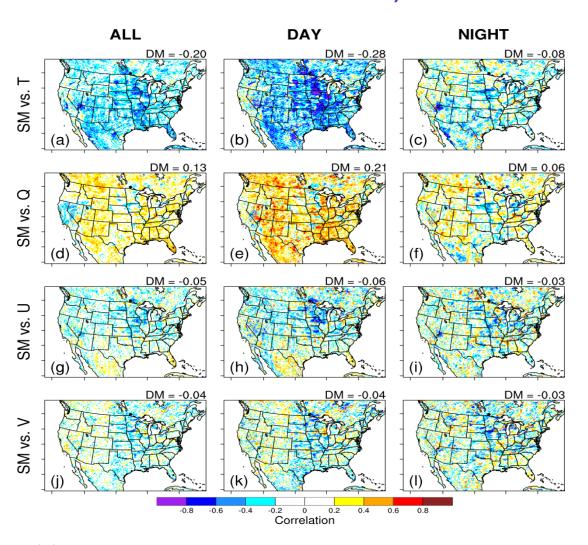
#### Sensitivity of near-surface variable forecasts to the changes in soil moisture and land use



#### Understanding covariances between soil and atmospheric states

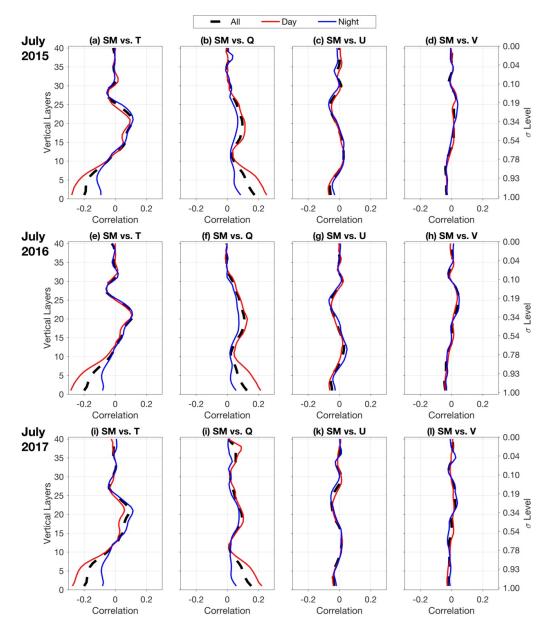
A strongly coupled land-atmospheric system (WRF-Noah)

L.-F. Lin and Z. Pu, JAMC 2018



"NMC-method"

The error correlations between top-layer soil moisture (SM) and bottom-layer atmospheric T, Q, U, and V in July 2016.



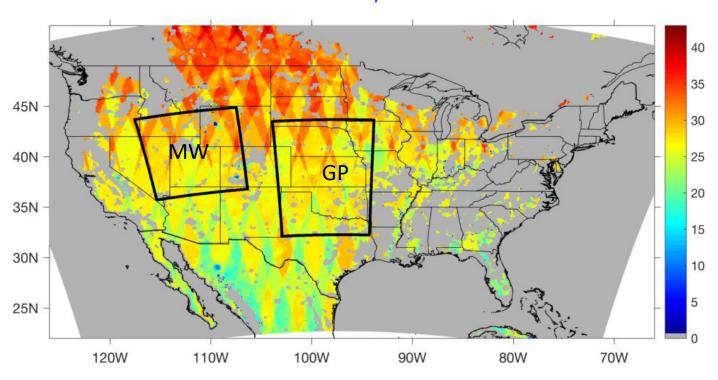
The domain mean correlation error between the top 10cm WRF-Noah soil moisture (SM) atmospheric states including potential (T), temperature specific humidity (Q), zonal wind (U), and meridional wind (V) in July from 2015 to 2017.

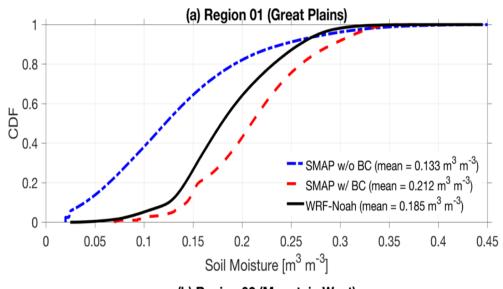
### Examining the influence of strong coupling on soil moisture data assimilation with SMAP satellite data: WRF-Noah

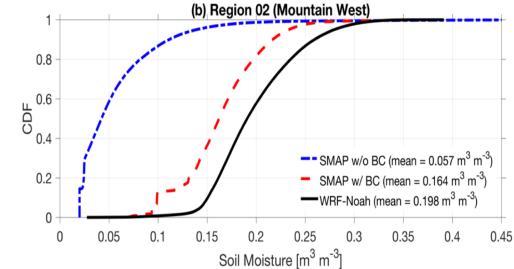
L.-F. Lin and Z. Pu, 2018 (preparation)

 Using Version 01 NASA SMAP 9-km enhancement soil moisture with quality control (removing data over surface types of vegetation, urban, water, and snow).

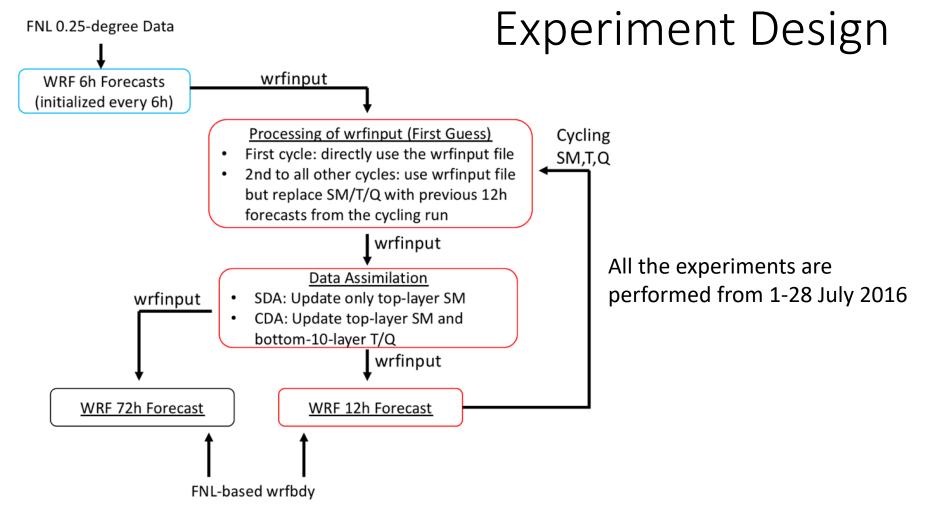
The sample of both descending and ascending data from SMAP in July 2016







- The soil moisture from Noah and SMAP SM before and after rescaling in July 2016 over the regions of interest
- Cumulative distribution function (CDF) matching



- OPL: no any data assimilation
- SDA: update only top-layer SM using non-bias-corrected SMAP SM
- CDA: update SM and T/Q using non-bias-corrected SMAP SM
- SDA\_SMBC: same as SDA, except using bias-corrected SMAP SM
- CDA\_SMBC: same as CDA, except using bias-corrected SMAP SM

#### **Evaluation Method and Reference Datasets**

- Reference datasets:
  - SM: SCAN and CRN gauges
  - T/Q: NCEP FNL 0.25-degree Analysis
- Evaluation Method:

Bias = 
$$\frac{1}{N} \sum_{i=1}^{N} (M_i - O_i)$$

RMSE = 
$$\sqrt{\frac{1}{N} \sum_{i=1}^{N} (M_i - O_i)^2}$$
,

$$\rho = \frac{\sum_{i=1}^{N} (M_i - \overline{M})(O_i - \overline{O})}{\sqrt{\sum_{i=1}^{N} (M_i - \overline{M})^2} \sqrt{\sum_{i=1}^{N} (O_i - \overline{O})^2}},$$

RI (Relative Improvement)

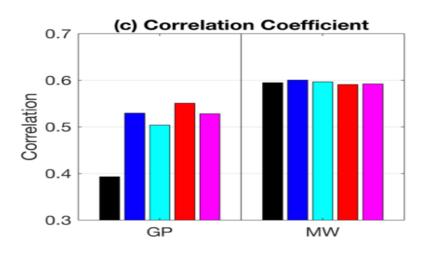
$$RI_{Bias} = \left(1 - \left| \frac{Bias_{DA}}{Bias_{OL}} \right| \right) \times 100\%$$

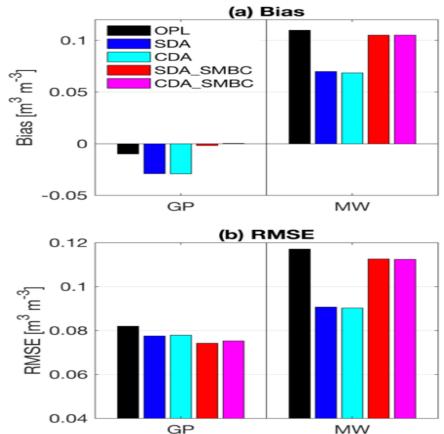
$$RI_{RMSE} = \frac{RMSE_{OL} - RMSE_{DA}}{RMSE_{OL}} \times 100\%$$

$$RI_{\rho} = \frac{\rho_{DA} - \rho_{OL}}{1 - \rho_{OL}} \times 100\%$$

#### **Top 10-cm SM Evaluation**

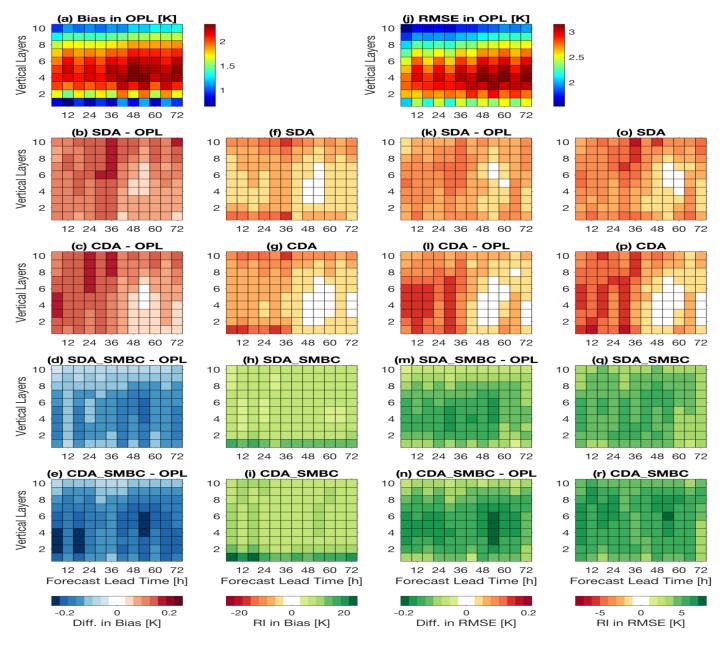
Averaged bias, RMSE, and correlation over the regions of interest.





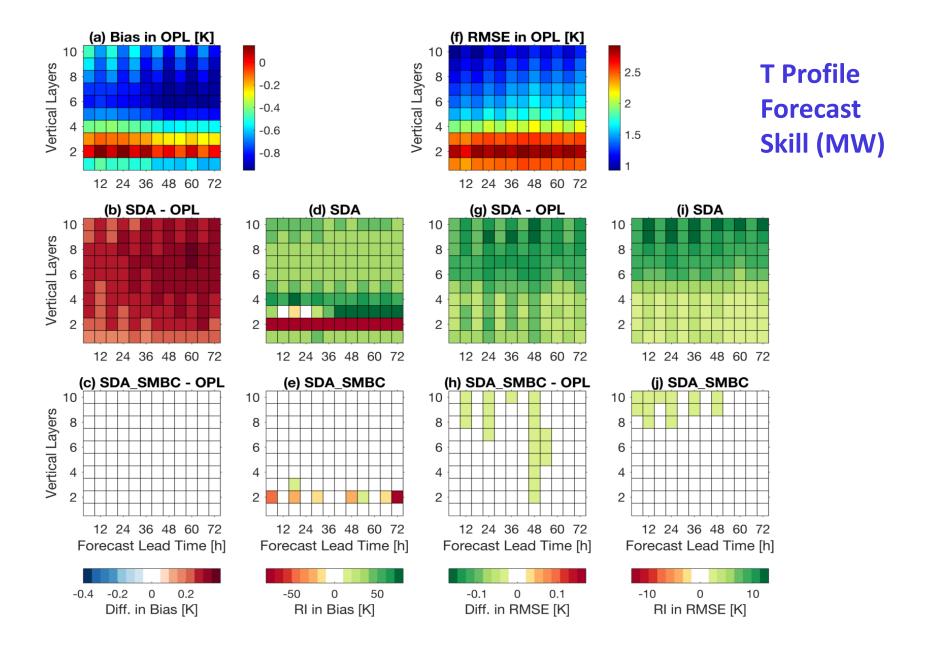
The relative improvement in terms of RMSE and correlation is reported in the table. A positive RI value means that DA improves the model skill.

	GP		MW	
	RMSE RI (%)	CORR RI (%)	RMSE RI (%)	CORR RI (%)
SDA	5	22	23	1
CDA	5	18	23	0
SDA_SMBC	9	25	4	-1
CDA_SMBC	8	22	4	-1



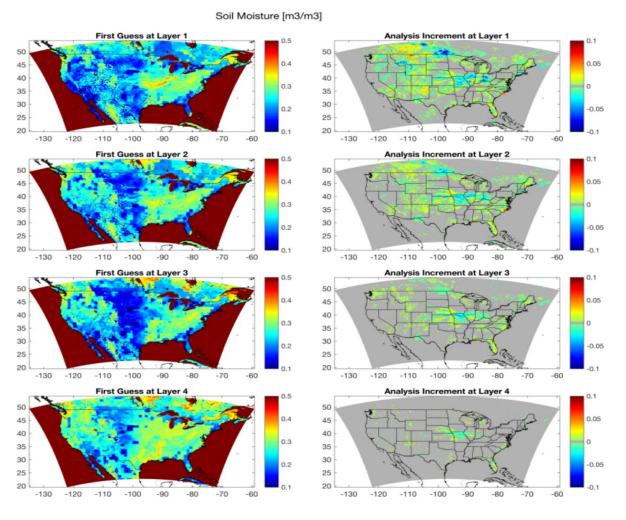
T Profile Forecast Skill (GP)

The domain mean metrics (bias and RMSE) over the Great Plains during 11-28 July 2016.



## Ongoing development I: Strong coupling within the GSI framework (Postdoc. Dr. L.-F. Lin)

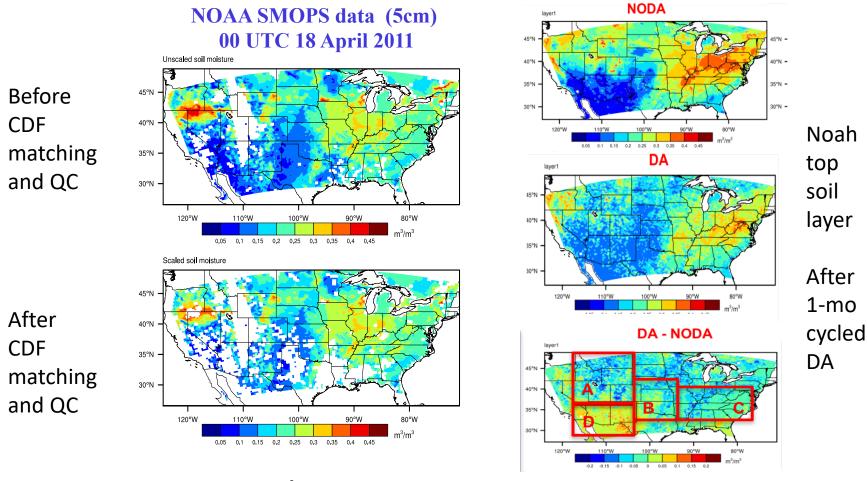
#### 00 UTC 4 July 2016 -- Soil moisture and analysis increment



To implement the state of soil moisture into GSI EnKF, so that assimilation of conventional data can have impacts on the soil moisture states.

#### On-going development II: NGGPS (NCEP FV3) coupling with NASA LIS

Brought up NASA Land Information System (LIS)



- Attended EMC/NCEP FV3 training course (June 2018)
- Plan to work on FV3-LIS coupling

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